Unsupervised Machine Learning For Identifying Signatures Of Past Mergers In Galaxy Light Profiles

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Background

- Galaxy interactions → stars and gas being pulled into different shapes (tidal features)
 - Streams & shells
- Difficult to detect by eye
- Solution: Machine learning!
 - Previous research: not enough labels for supervised → try unsupervised



Fig. 1. Two galaxies interacting, producing tidal streams. Photo credit: HSC-SSP

1. Scale images

<u>Image Dataset</u>: Hyper Suprime-Cam Subaru Strategic Program PDR-3 512x512 pixels in g,r,i bands with labels for tidal features (for method testing)



2. Augmentations of scaled images

- Maintain tidal features
- Result: image pair
 - one image scaled and augmented, one scaled and non-augmented
- 3. Convolution Step



- 3. Compute similarity (loss)
- Between augmented and non-augmented image pair

$$ext{Cosine Similarity} = rac{\sum_{i=1}^n x_i y_i}{\sqrt{\sum_{i=1}^n x_i^2} \sqrt{\sum_{i=1}^n y_i^2}}$$

5. Adjust weights through iteration

<u>Goal</u>: minimize loss function between images within pairs and maximize it between different pairs

Output

multidimensional representation of dataset

- Similar images grouped
- Scale to 2D
- Plot parameters



Results

Top Plots

 Clear trends in galaxy redshift and galaxy size





Bottom plots

- Faint trend in tidal shell plot
- No trend in tidal streams

Results

Run again

- Each image has its reflection subtracted
- Emphasis on asymmetric data like tidal features
- Improved results



Conclusion

- Unsupervised machine learning does separate galaxies based on their parameters, but not the parameters we want
- To continue improving the model, we need a larger set of galaxy images and to experiment with augmentations.









Thank you!



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Streams

Shells

